Getting Ready

Content, community, structure, when to write, plan

Ming-Ming Cheng

College of Computer Science, Nankai University

Course materials: https://mmcheng.net/writing/

2022/05/24

Table of Contents

- 1 Content: novelty centered
- Novelty types: new problem, new approach, combination innovation
 - New problem
 - New approach
 - Combination innovation
- 3 Community and where to publish
- 4 Paper structure and writing plan



Fig.: Sections 1-2.

Conditions for a good publication

Content + Style

- your great idea
- your advance in the subject.

Relationship between content and style

- Without content, style is irrelevant
- Even with a major break-through, your paper may still get rejected. *e.g.* The TIP example.
- High quality writing \rightarrow the chance of acceptance.

Why do we need a paper?

Readers are interested in your valuable discovery or take-home message, rather than your showoff.

Key: novelty and originality

Originality in decreasing order \downarrow

- Find a new problem to solve.
- New approach to solve an existing problem.
- Put existing ideas together in a novel way, to solve some known problem.

Content

Your method should be an improvement in some way.

- Faster
- More accurate
- Less memory consumption
-

Chose the way you wish to claim!

Exception

Negative results can also represent advances in knowledge. If they are unexpected.

- Most people familiar with the topic would expected to be better.
- But you have shown for some unforeseen reason that it is actually worse.

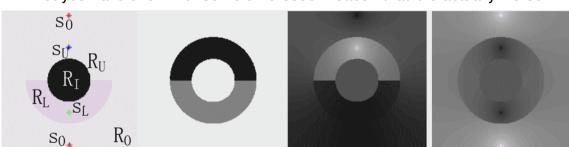


Fig.: Connectedness of Random Walk Segmentation, IEEE TPAMI 2011.

Put just one good idea in each paper!

- Even your main idea is brilliant, a reviewer can recommend rejection since you second idea is weak!
- You may not be able to fully describe, justify, and provide experimental supporting to your idea.
- I typically claim no more than 2 contributions.

The problem must be

- sufficiently difficult, i.e. challenging.
- of use or interest to potential readers.

Finding a good new problem is tricky. e.g.

- Find human faces: not novel.
- Find snakes: probably too difficult, not very useful, too specific.



Fig.: Sketch2Photo: Internet Image Montage, ACM TOG 2009.

11/39



Fig.: ImageSpirit: Verbal Guided Image Parsing, ACM TOG 2014.

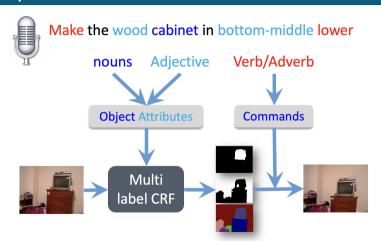


Fig.: ImageSpirit: Verbal Guided Image Parsing, ACM TOG 2014.

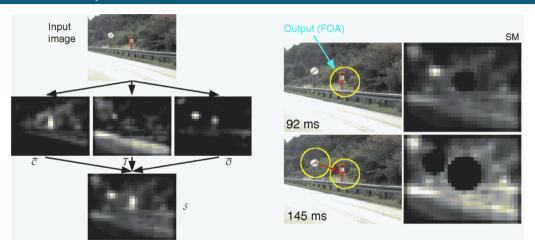


Fig.: A model of saliency-based visual attention for rapid scene analysis, IEEE TPAMI 1998.

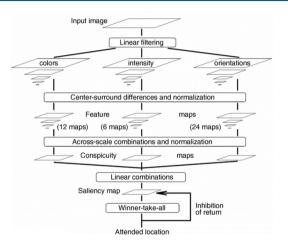


Fig.: A model of saliency-based visual attention for rapid scene analysis, IEEE TPAMI 1998.

15 / 39

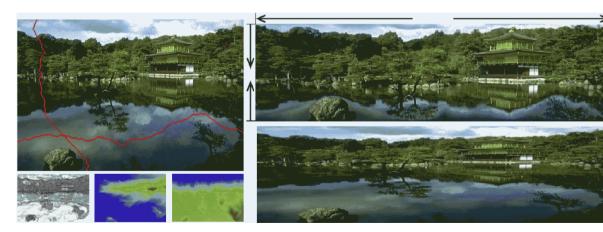


Fig.: Seam Carving for Content-Aware Image Resizing, ACM TOG 2007.

New approach to solve

- an existing problem.
- some step in a problem.



Fig.: Shape-Preserving Approach to Image Resizing, PG 2009.

18 / 39

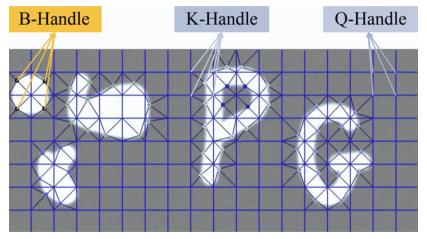


Fig.: Shape-Preserving Approach to Image Resizing, PG 2009.

Cited by 11314 Related articles All 45 versions

A model of saliency-based **visual attention** for **rapid** scene analysis

<u>L Itti, C Koch, E Niebur</u> - IEEE Transactions on pattern analysis ..., 1998 - ieeexplore.ieee.org

A **visual attention** system, inspired by the behavior and the neuronal architecture of the early primate **visual** system, is presented. Multiscale image features are combined into a single topographical saliency map. A dynamical neural network then selects attended locations in ...

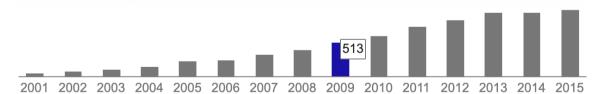


Fig.: A model of saliency-based visual attention for rapid scene analysis, IEEE TPAMI 1998.

```
%% Read image from file
inImg = im2double(rgb2gray(imread('yourImage.jpg')));
inImg = imresize(inImg, 64/size(inImg, 2));
%% Spectral Residual
mvFFT = fft2(inImg):
mvLogAmplitude = log(abs(mvFFT));
mvPhase = angle(mvFFT);
mySpectralResidual = myLogAmplitude - imfilter(myLogAmplitude, fspecial('average',
saliencvMap = abs(ifft2(exp(mvSpectralResidual + i*mvPhase))).^2;
%% After Effect
saliencvMap = mat2grav(imfilter(saliencvMap, fspecial('gaussian', [10, 10], 2.5)));
imshow(saliencvMap);
```

Fig.: Saliency Detection: A Spectral Residual Approach, IEEE CVPR 2007.

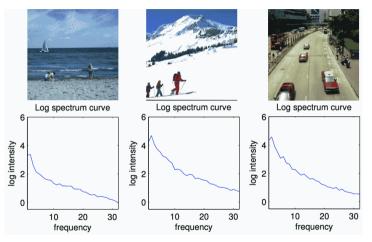
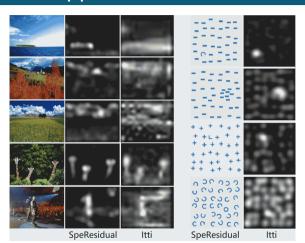


Fig.: Saliency Detection: A Spectral Residual Approach, IEEE CVPR 2007.



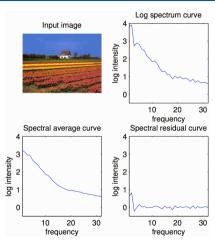


Fig.: Saliency Detection: A Spectral Residual Approach, IEEE CVPR 2007.

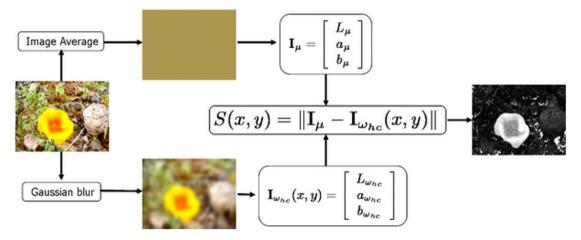


Fig.: Frequency-tuned Salient Region Detection, IEEE CVPR 2009.

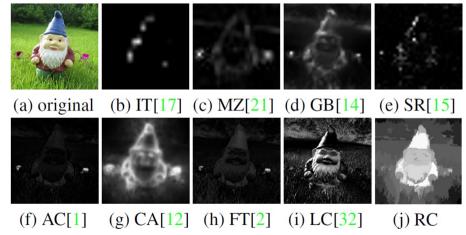
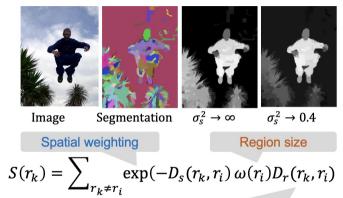


Fig.: Global Contrast based Salient Region Detection, IEEE TPAMI 2015 (CVPR 2011).



Region contrast by sparse histogram comparison.

Fig.: Global Contrast based Salient Region Detection, IEEE TPAMI 2015 (CVPR 2011).

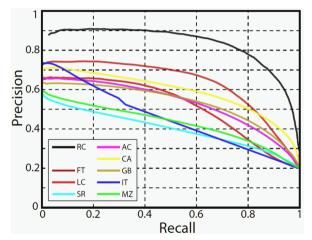


Fig.: Global Contrast based Salient Region Detection, IEEE TPAMI 2015 (CVPR 2011).

The originality here would be

- the choice of suitable components.
- the way of linking these components together to meet an overall goal.

Should have some clever insight!

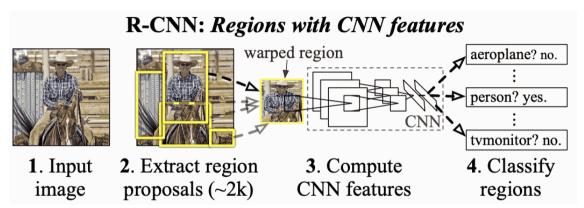


Fig.: Rich feature hierarchies for accurate object detection and semantic segmentation, IEEE CVPR 2014.

Rich feature hierarchies for accurate object detection and semantic segmentation, IEEE CVPR 2014.

- To what extent do the CNN classification results on ImageNet generalize to object detection results on the PASCAL VOC Challenge?
- 2K category-independent region proposals → CNNs
- supervised pre-training on a large auxiliary dataset (ILSVRC), followed by domain- specific fine-tuning on a small dataset (PASCAL)

Very useful insights!

VOC 2010 test	aero	bike	bird	boat	bottle	bus		sofa	train	tv	mAP
DPM v5 [18] [†]	49.2	53.8	13.1	15.3	35.5	53.4	 	20.7	43.8	38.3	33.4
UVA [34]	56.2	42.4	15.3	12.6	21.8	49.3		31.8	47.0	44.8	35.1
Regionlets [36]	65.0	48.9	25.9	24.6	24.5	56.1		32.6	54.0	45.9	39.7
SegDPM [16] [†]	61.4	53.4	25.6	25.2	35.5	51.7		35.0	52.8	43.1	40.4
R-CNN	67.1	64.1	46.7	32.0	30.5	56.4		38.1	52.8	50.2	50.2
R-CNN BB	71.8	65.8	53.0	36.8	35.9	59.7		39.3	61.2	52.4	53.7

Fig.: Rich feature hierarchies for accurate object detection and semantic segmentation, IEEE CVPR 2014.

Community and where to publish

Different fields have different unstated rules and expectations about:

- How a paper should be written?
- How to describe approaches?
- How to analyze results?
- . . .

E.g. computer graphics, computer vision, or image processing.

Community and where to publish

IEEE TPAMI

The journal paper be a "substantial revision" (>30 percent) of the conference publication. Examples of the improvements: additional technical details, a clearer explanation of the contribution, more experiments if appropriate, or an updated state-of-the-art.

Community 0000

IFFF TIP

Clearly identifiable benefit that offers to the research community beyond the already published conference paper. e.g. additional analysis, novel algorithmic enhancements, added theoretical work, completeness of exposition, extensive experimental validation, etc.

Follow the expectations/style of a community

- Computational geometry paper typically have a theoretical analysis of how the performance scales with the quantity of input data, yet not practically testing their algorithms.
- Computer graphics papers emphasis on practical testing and quality of output.
- Computer vision paper typically pay more attention to benchmark results.

. . .

Graphics: SIGGRAPH ← ACM TOG

CV: IEEE CVPR → IEEE TPAMI

CS community is special!

0000

- Evaluate relative quality of journals: impact factor, h5-index.
- Journals are less urgency, more likely to have reversion, more opportunity.

Employers assessing your research ability will often care where you publish more than your work itself!

Abstract and paper

- Publishers usually freely provide abstracts, wishing readers to pay before accessing the full paper.
- The abstract forms a standalone, separate, small document.
- The paper should not omit anything that is said in the abstract.

Paper structure

Header:

- Title
- Author list
- Author's affiliations and address
- Abstract
- Keywords

Paper structure

Body of paper:

- Introduction
- Related works
- Algorithms or Methods or Theoretical results
- Experiments or Proofs
- Conclusions (including possible future work)

Writing plan

- Write first draft fairly quickly and then revise it.
- Know what has already been explained earlier in the paper.
- Know what is still left to explain.
- Avoid write extensive detail and then find there is no space for it.
- Writing forces you to rethink about experiments: redo experiments or modify an algorithm.
- Cover everything that need saying, in enough detail, and then polish and shorten it during reversion.